

week - 3

7/18/25

aim : to compare different classification algorithm using a open-source dataset based on accuracy precision recall and f1 score

procedure:

- 1, import the required libraries
- 2, load the iris dataset (built-in)
- 3, split the dataset into training and test sets
- 4, train two classifiers:
 - logistic regression
 - k-Nearest Neighbours
- 5, use both models to make predictions
- 6, evaluate each model using:
 - accuracy
 - classification report
- 7, print and compare the result

code:

```
from sklearn.datasets import load_iris  
from sklearn.model_selection import train_test_split  
from sklearn.linear_model import LogisticRegression  
from sklearn.neighbors import KNeighborsClassifier  
from sklearn.metrics import accuracy_score,  
                           classification_report  
  
iris = load_iris()  
X = iris.data  
y = iris.target  
  
X_train, X_test, y_train, y_test = train_test_split(  
    X, y, test_size=0.3, random_state=42)  
lr = LogisticRegression(max_iter=200)  
lr.fit(X_train, y_train)
```

$lr_pred = lr.predict(x_test)$

print ("logistic Regression: ")

print ("accuracy : ", accuracy_score(y-test, lr_pred))

print (classification_report (y-test, lr_pred))

knn = KNeighborsClassifier(n_neighbors=3)

knn.fit (x-train, y-train)

knn_pred = knn.predict (x-test)

print ("In k-nearest Neighbors: ")

print ("accuracy : ", accuracy_score(y-test, knn_pred))

print (classification_report (y-test, knn_pred))

Output: logistic Regression :-

Accuracy : 0.977

@	precision	recall	F1Score	Support
0	1.00	1.00	1.00	16
1	1.00	0.94	0.97	18
2	0.92	1.00	0.96	11
accuracy			0.98	45
macro avg	0.97	0.98	0.98	45

K-nearest Neighbors:-

Accuracy : 0.9777

@	precision	recall	F1 Score	Support
0	1.00	1.00	1.00	16
1	1.00	0.94	0.97	18
2	0.92	1.00	0.96	11
accuracy			0.98	45
macro avg	0.97	0.98	0.98	45

$$\text{Accuracy} = \frac{\text{total positive} + \text{total negative}}{\text{TP} + \text{TN} + \text{FP} + \text{FN}}$$

$$\text{Precision} = \frac{\text{TP}}{\text{TP} + \text{FP}} \quad (\text{How many are actual positive})$$

$$\text{Recall} = \frac{\text{TP}}{\text{TP} + \text{FN}} \quad (\text{How many actual positive})$$

F1 score = Harmonic mean of precision and recall

Result: in this experiment I have successfully implemented the combining of different classification algorithms using open source dataset.

```

import numpy as np
import pandas as pd
from sklearn.datasets import load_iris
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import StandardScaler
from sklearn.metrics import accuracy_score, precision_score, recall_score, f1_score, confusion_matrix
from sklearn.linear_model import LogisticRegression
from sklearn.tree import DecisionTreeClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.neighbors import KNeighborsClassifier
from sklearn.svm import SVC
from sklearn.naive_bayes import GaussianNB

iris = load_iris()
X = iris.data
y = iris.target

X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size=0.3, random_state=42
)

scaler = StandardScaler()
X_train = scaler.fit_transform(X_train)
X_test = scaler.transform(X_test)
models = {
    "Logistic Regression": LogisticRegression(max_iter=200),
    "Decision Tree": DecisionTreeClassifier(),
    "Random Forest": RandomForestClassifier(),
    "K-Nearest Neighbors": KNeighborsClassifier(),
    "Support Vector Machine": SVC(),
    "Naive Bayes": GaussianNB()
}

results = []
for name, model in models.items():
    model.fit(X_train, y_train)
    y_pred = model.predict(X_test)

    acc = accuracy_score(y_test, y_pred)
    prec = precision_score(y_test, y_pred, average='weighted')
    rec = recall_score(y_test, y_pred, average='weighted')
    f1 = f1_score(y_test, y_pred, average='weighted')
    cm = confusion_matrix(y_test, y_pred)

    results.append([name, acc, prec, rec, f1])

print(f"\n{name}")
print("*" * 40)
print("Confusion Matrix:\n", cm)
print(f"Accuracy : {acc:.4f}")
print(f"Precision: {prec:.4f}")
print(f"Recall   : {rec:.4f}")
print(f"F1-Score : {f1:.4f}")

df = pd.DataFrame(results, columns=["Classifier", "Accuracy", "Precision", "Recall", "F1-Score"])
print("\n\nSummary of Classifier Performance:")
print(df.to_string(index=False))

```

Logistic Regression
=====

Confusion Matrix:
[[19 0 0]
 [0 13 0]
 [0 0 13]]
Accuracy : 1.0000
Precision: 1.0000
Recall : 1.0000
F1-Score : 1.0000

Decision Tree
=====

Confusion Matrix:
[[19 0 0]
 [0 13 0]]

```
[ 0  0 13]]
Accuracy : 1.0000
Precision: 1.0000
Recall   : 1.0000
F1-Score : 1.0000

Random Forest
=====
Confusion Matrix:
[[19  0  0]
 [ 0 13  0]
 [ 0  0 13]]
Accuracy : 1.0000
Precision: 1.0000
Recall   : 1.0000
F1-Score : 1.0000

K-Nearest Neighbors
=====
Confusion Matrix:
[[19  0  0]
 [ 0 13  0]
 [ 0  0 13]]
Accuracy : 1.0000
Precision: 1.0000
Recall   : 1.0000
F1-Score : 1.0000

Support Vector Machine
=====
Confusion Matrix:
[[19  0  0]
 [ 0 13  0]
 [ 0  0 13]]
Accuracy : 1.0000
Precision: 1.0000
Recall   : 1.0000
F1-Score : 1.0000

Naive Bayes
=====
```